# CS368 MATLAB Programming Lecture 7

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Based on lecture slides by Michael O'Neill and Beck Hasti

March 9, 2022

# Coordination Game

- No lecture next week.
- For the quiz grade next week (out of 2):
- A: If  $\geq 95\%$  of you chooses A, everyone gets 3 points.
- $B: If \ge 75\%$  of you chooses B, everyone gets 2 points.
- C: If  $\geq 50\%$  of you chooses C, everyone gets 1 point.
- Otherwise, everyone gets 0.

#### Schedule Admin

- No lecture next week.
- There will be office hours (all on Zoom).
- Code similarity check. Grade maybe changed to 0 if:
- No code submission on Canvas (either .txt or .m). Allowed to resubmit.
- ② Almost identical code without attribution. Not allowed to resubmit.
  - Midterm grades on Canvas. Not going to submit official midterm grades.

#### Polynomials Math

- A degree n polynomial,  $c_n x^n + c_{n-1} x^{n-1} + c_{n-2} x^{n-2} + ... + c_2 x^2 + c_1 x + c_0$  is specified by a list of coefficients  $c_n, c_{n-1}, ..., c_1, c_0$  usually stored in a vector.
- The computation can be done efficiently using Horner's method:  $((((c_nx + c_{n-1})x + c_{n-2})x + ... + c_2)x + c_1)x + c_0$ .

# Exponential and Trigonometry Functions

• Many elementary function can be evaluated and approximated using polynomials.

• 
$$e^x \approx 1 + x + \frac{x^2}{2} + \frac{x^3}{6} + \frac{x^4}{24} + \frac{x^5}{120} + \dots$$

$$\circ$$
  $\sin(x) \approx x - \frac{x^3}{6} + \frac{x^5}{120} + \dots$ 

$$\cos(x) \approx 1 - \frac{x^2}{2} + \frac{x^4}{24} + \dots$$

• They are also called Taylor polynomials.

# Polynomial Evaluation

- Suppose c and x are row vectors with length n.
- $dot(c, \times .^{\hat{}} ((n-1):-1:0))$  evaluates the polynomial with coefficients c at x.
- polyval(c, x) evaluates the same polynomial but faster.

# Polynomial Evaluation, Trig

• (Compute 
$$\cos(x) \approx 1 - \frac{x^2}{2} + \frac{x^4}{24} + \dots - \frac{x^{10}}{10!}$$
 at  $x = 1$ .)

- 0.5403
- C: polyval(((-1) .^ (5:-1:0)) ./ (factorial (10:-2:0)), 1)

#### Lagrange Polynomial

- A set of n points (with distinct x values) uniquely determine a degree n-1 polynomial that passes through all the points.
- The coefficients of the leading terms can be zero, so technically the degree of the polynomial can be less than n-1.

# Lagrange Polynomial Diagram Math

# Vandermonde Matrix

The coefficients can be found by solving a system of linear

equations, 
$$\begin{bmatrix} x_1^{n-1} & x_1^{n-2} & \dots & 1 \\ x_2^{n-1} & x_2^{n-2} & \dots & 1 \\ \dots & \dots & \dots & \dots \\ x_n^{n-1} & x_n^{n-2} & \dots & 1 \end{bmatrix} \begin{bmatrix} c_{n-1} \\ c_{n-2} \\ \dots \\ c_0 \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \dots \\ y_n \end{bmatrix}.$$

## Polyfit Code

- Suppose x, y are two row vectors with length n.
- repmat(x', 1, n) . ((n 1):-1:0) is the Vandermonde matrix, and vander(x) creates the same matrix.
- vander(x) \ y' solves for the coefficients using the Vandermonde matrix.
- polyfit (x, y, n 1) finds the same coefficients.

### Spline Math

- A set of n points (with distinct x values, and ordered according to x) can define n-1 cubic polynomial segments satisfying the following conditions.
- **1** The *i*-th polynomial segment connects point i and point i + 1.
- ② The segments form a smooth curve. Technically,  $C^2$  smoothness is desired: the first and second derivatives should be continuous.

# Spline Diagram Math

# Tridiagonal Matrix

• With n points, there are 4n - 4 coefficients to compute. Suppose the coefficient for segment i connecting  $(x_i, y_i)$  and

 $(x_{i+1}, y_{i+1})$  is  $c_{i1}, c_{i2}, c_{i3}, c_{i4}$ 

- $c_{i1}x_i^3 + c_{i2}x_i^2 + c_{i3}x_i + c_{i4} = y_i$  for i = 1, 2, ..., n 1 so that the segment passes through the left end point.
- ②  $c_{i1}x_{i+1}^3 + c_{i2}x_{i+1}^2 + c_{i3}x_{i+1} + c_{i4} = y_{i+1}$  for i = 1, 2, ..., n-1 so that the segment passes through the right end point.
- 3  $3c_{i,1}x_{i+1}^2 + 2c_{i2}x_{i+1} + c_{i3} = 3c_{i+1,1}x_{i+1}^2 + 2c_{i+1,2}x_{i+1} + c_{i+1,3}$  for i = 1, 2, ..., n-2 so that the first derivative is continuous.
- **3**  $6c_{i1}x_{i+1} + 2c_{i2} = 6c_{i+1,1}x_{i+1} + 2c_{i+1,1}$  for i = 1, 2, ..., n-2 so that the second derivative is continuous.
  - There are 4n 6 equations. 2 more equations are needed.

## End Point Conditions

- There are many ways to specify the 2 extra equations.
- **1** If the third derivative for the first and last two segments are continuous,  $c_{1,1}=c_{2,1}$  and  $c_{n-1,1}=c\ (n-2,1)$ , the resulting spline is called a not-a-knot spline.
- ② If the second derivative at the end points are 0, or  $6c_{1,1}x_1 + 2c_{1,2} = 0$  and  $6c_{n-1,1}x_n + 2c_{n-1,2} = 0$ , the resulting spline is called a natural spline.

# Spline

- Suppose x, y are two row vectors with length n.
- spline (x, y). coefs finds the coefficients of the cubic spline that interpolates (x, y) while satisfying the not-a-knot condition.
- spline (x, y, x0) evaluates the cubic spline interpolating (x, y) at  $x = x_0$ .
- csape(x, y, 'variational'). coefs finds the coefficients of the natural cubic spline. It requires MATLAB's Curve Fitting Toolbox.

#### Local Coefficient

- spline (x, y). coefs are local coefficients meaning the equation for segment i and  $x \in [x_i, x_{i+1}]$  is  $c_{i1}(x x_i)^3 + c_{i2}(x x_i)^2 + c_{i3}(x x_i) + c_{i4}$ .
- The polynomial can be expanded to find the coefficients of the original spline problem.

# Interpolation, Polyfit

```
1 x = 1:5; y = polyval(ones(5, 1), x); polyfit(x, y, 4)
```

• B:1 1 1 1 1

• C: 0 0 0 0 1

# Interpolation, Polyfit Too

- **1** x = 1:5; y = 5:-1:1; polyval(polyfit(x, y, 4), x)
  - C:5 4 3 2 1
  - D:1 2 3 4 5

# Interpolation, Spline Quiz

- **1** x = 1:5; y = 5:-1:1; spline(x, y, x)
  - C:5 4 3 2 1
  - D:1 2 3 4 5

# Interpolation, Spline Too

• 
$$C: \frac{3}{2}$$

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#### Polynomial Regression

#### Math

- polyfit (x, y, d) with d < n, finds the polynomial that is the closest to the points (x, y).
- The total distance from the points to the polynomial is defined as the sum of the squared differences between the y value at x and the value of the polynomial at x,

$$D = \sum_{i=1}^{n} \left( y_i - \sum_{j=0}^{d} c_i x_i^j \right)^2$$
$$= \sum_{i=1}^{n} \left( y_i - c_d x_i^d - c_{d-1} x_i^{d-1} - \dots - c_1 x_i - c_0 \right)^2.$$

polyfit finds the coefficients of the polynomial that minimizes
 D. These coefficients are also called regression coefficients.

# Regression Diagram Math

# General Linear Regression

- regress(y', X) finds the regression coefficients for  $c_nx_n + c_{n-1}x_{n-1} + ... + c_1x_1 + c_0x_0$ , where  $x_0, x_1, ..., x_n$  are the columns of X. It requires MATLAB's Statistics and Machine Learning Toolbox.
- regress (y', X) solves  $(X^TX) c = X^Ty$  instead of Xc = y. This is because  $X^TX$  is always invertible (meaning  $(X^TX)^{-1}$  always exists), so  $c = (X^TX)^{-1}X^Ty$ .
- When X is the degree d Vandermonde matrix for x, meaning  $x_i = x^i$  for i = d, d 1, ..., 0, regress (y, X) finds the same coefficients as polyfit (x, y, d).

## Non-Polynomial Regression

- polyfit (f(x), y, 1) finds the regression coefficients for  $c_1f(x) + c_0$ , for example, polyfit  $(\sin(x), y, 1)$  finds  $c_1, c_0$  such that the total distance from the points to the function  $y = c_1 \sin(x) + c_0$  is minimized.
- regress (y', [f1(x') f2(x') ...]) finds the regression coefficients for  $c_1f_1(x) + c_2f_2(x) + ...$ , for example, regress (y', [sin(x') cos(x') ones(length(x), 1)]) finds  $c_2, c_1, c_0$  such that the total distance from the points to the function  $y = c_2 \sin(x) + c_1 \cos(x) + c_0$  is minimized.

# Regression, Polynomial

- (The values of c, d such that the curve  $y = cx^2 + d$  best fits (x, y).)
- 1
- - A: polyfit (x, y, 1)
  - B: polyfit (x, y, 2)
  - C: polyfit (x .^ 2, y, 1)

# Regression, Unknown Degree

- (The values of c, d such that the curve  $y = cx^d$  best fits (x, y).)
- 1
- - A: v = polyfit(log(x), log(y), 1); [exp(v(2)) v(1)]
  - B: v = polyfit(x, log(y), 1); [exp(v(2)) v(1)]
- C: v = polyfit(log(x), log(y), 1); [v(2) v(1)]
- D: v = polyfit(x, log(y), 1); [v(2) v(1)]

# Regression, Non-Polynomial

- (The values of c, d such that the curve  $y = ce^{dx}$  best fits (x, y).)
- 1
- **1** x = 1:10;  $y = \exp(x)$ ;
  - A: v = polyfit(log(x), log(y), 1); [exp(v(2)) v(1)]
  - B: v = polyfit(x, log(y), 1); [exp(v(2)) v(1)]
- C: v = polyfit(log(x), log(y), 1); [v(2) v(1)]
- D: v = polyfit(x, log(y), 1); [v(2) v(1)]

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